

The_final_code

Anders Hjulmand, Morten Gade, Sigrid Nielsen and Gustav Helms

2/3/2020

```
#Setup
library(pacman)
pacman::p_load("tidyverse")

#Loading the data
df <- read.csv("sleepstudy.csv", header = T)
df$Subject <- as.factor(df$Subject)
```

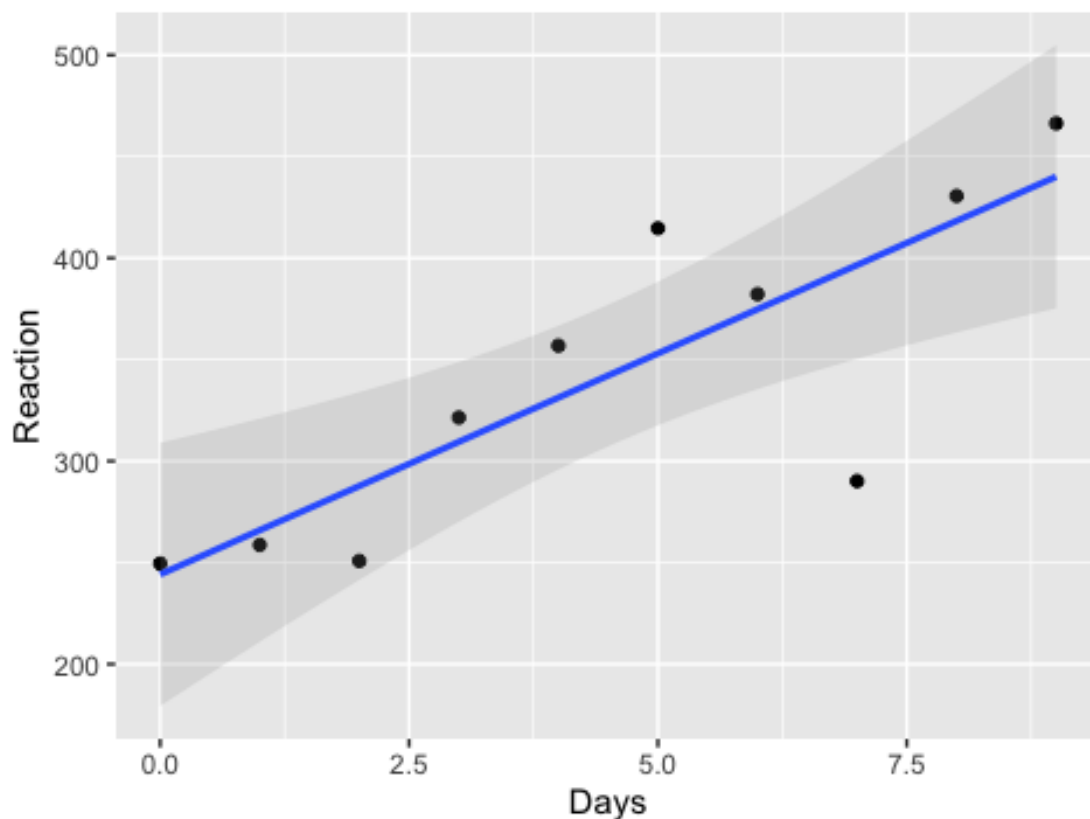
Task 1

1.a: Get the data from one participant, e.g. using subset(). Make a linear regression for reaction time as a function of days of sleep deprivation, e.g. using lm(). Report the F-statistics.

```
#Making a subset df
participan1 <- subset(df, df$Subject == 308)

#Making a scatterplot of subject 308
ggplot(participan1, aes(Days, Reaction))+
  geom_point()+
  geom_smooth(method = lm, alpha = 0.2)+
  ggtitle("Scatterplot of subject 308's reaction time by days of sleep deprivation")
```

Scatterplot of subject 308's reaction time by days of sle



```
#Linear regression model
summary(lm(participan1$Reaction~participan1$Days))

##
## Call:
## lm(formula = participan1$Reaction ~ participan1$Days)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -106.397   -4.098    9.688   22.269   61.674
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      244.19      28.08   8.695 2.39e-05 ***
## participan1$Days      21.77       5.26   4.137 0.00326 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 47.78 on 8 degrees of freedom
## Multiple R-squared:  0.6815, Adjusted R-squared:  0.6417
## F-statistic: 17.12 on 1 and 8 DF, p-value: 0.003265
```

A linear regression analysis was used to test if the amount of days of sleep deprivation significantly predicted reaction time. The results of the regression indicated that the predictor explained 64.17% of the variance in the reaction time (Adjusted R² = .6417, F(1,8) = 17.12, p < .01). It was found that a larger duration of sleep deprivation significantly predicted reaction time tendencies ($\beta = 21.77$, SE = 5.26, t = 4.137, p < .01)

1.b: How many degrees of freedom does the relevant F-distribution have?

The relevant F-distribution have df=1 for predictors and df=8 for subjects.

1.c: At which F-value does a regression with this distribution become statistically significant (p<0.05)?

```
#Getting the F-score for the 95% significance cutoff using the quantile function
qf()
cutoff<-qf(0.95,1,8)
'p=0.05 cutoff value with df(1,8)'

## [1] "p=0.05 cutoff value with df(1,8)"

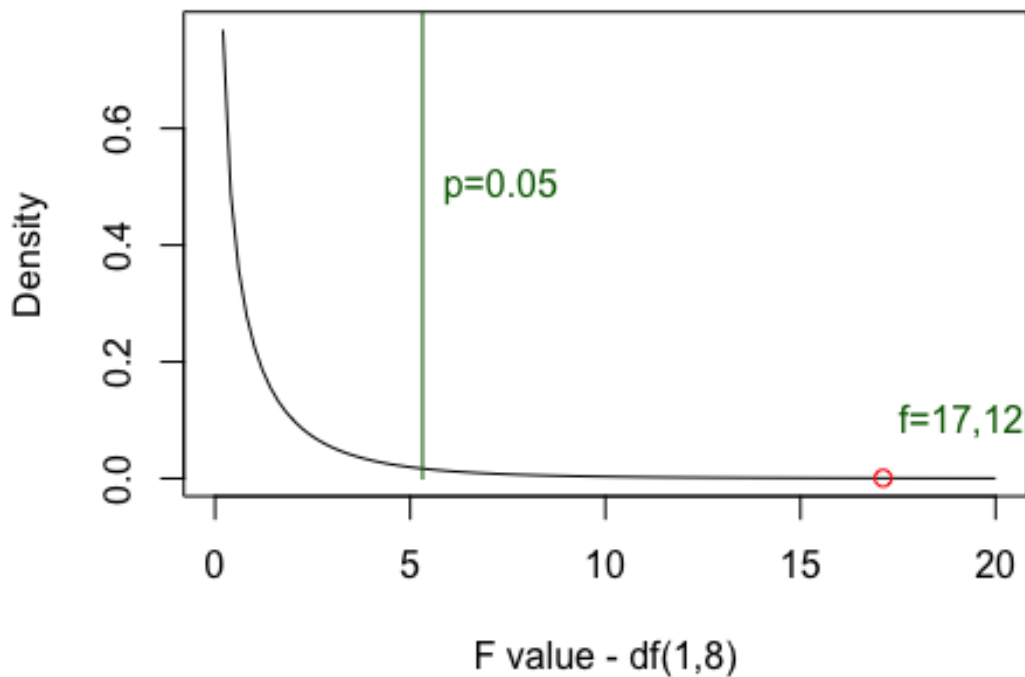
cutoff

## [1] 5.317655
```

A F-distribution with df(1,8) becomes statistically significant when F > 5.318 at a significance level of 0.05.

1.d: Make a plot of the F-distribution.

```
#The F statistics for the effect of days on reaction time
model<-lm(participant1$Reaction~participant1$Days)
res=anova(model)
#Making a string of 100 numbers between 0 and 20 to use when finding the F
distribution
nn<-seq(0,20,len=100)
#Getting the F distribution using df()
fdist<-df(nn,1,8)
plot(nn,fdist, type='l',xlab='F value - df(1,8)',ylab='Density')
#plotting a vertical line at the cutoff
lines(c(cutoff,cutoff),c(0,1),col='darkgreen')
#Add explanation for the line
text(cutoff+2,0.5,'p=0.05',col='darkgreen')
#draw F-value as point on the curve
points(res$F value[1],df(res$F value[1],1,8),col='red')
#add text for the f-value
text(res$F value[1]+2,0.1,'f=17,12',col='darkgreen')
```



Task 2. For all participant in the experiment

2.a: Find the coefficients (slope and intercept) for the regression for reaction time as a function of days of sleep deprivation (a hint for the solution: use `group_by()` in tidyverse or this function here: <https://stat.ethz.ch/R-manual/R-devel/library/nlme/html/lmList.html>, hint2: `pool=FALSE`)

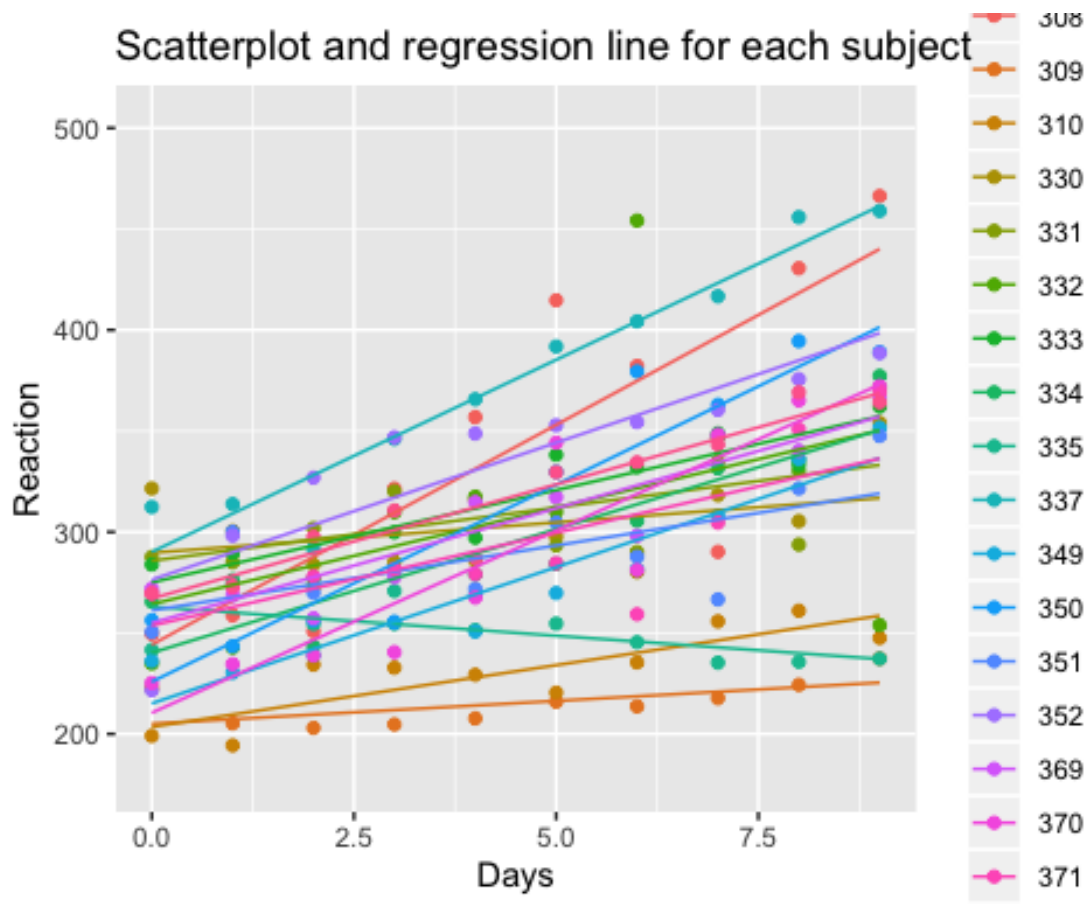
```
all_subjects <- df %>% group_by(Subject) %>% summarise("Intercept" =
lm(Reaction~Days)$coefficients[1],
                                                    "Slope" =
lm(Reaction~Days)$coefficients[2])
all_subjects
```

```
## # A tibble: 18 x 3
##   Subject Intercept Slope
##   <fct>      <dbl> <dbl>
## 1 308         244.  21.8
## 2 309         205.   2.26
## 3 310         203.   6.11
## 4 330         290.   3.01
## 5 331         286.   5.27
```

```
## 6 332      264.  9.57
## 7 333      275.  9.14
## 8 334      240. 12.3
## 9 335      263. -2.88
## 10 337     290. 19.0
## 11 349     215. 13.5
## 12 350     226. 19.5
## 13 351     261.  6.43
## 14 352     276. 13.6
## 15 369     255. 11.3
## 16 370     210. 18.1
## 17 371     254.  9.19
## 18 372     267. 11.3
```

2.b: Combine both scatter plot and regression line in the same figure. You may also include all participants in one plot.

```
#Making the scatterplot
ggplot(df, aes(Days, Reaction, colour = Subject))+
  geom_point()+
  geom_smooth(method = lm, alpha = 0, size = 0.5)+
  ggtitle("Scatterplot and regression line for each subject")
```



2.c: Collect and report the inferential statistics for each participant in a table using t-statistics, including t-value, df and p-value.

2.d: How many individual participants display a statistically significant effect of sleep deprivation (p-values uncorrected for multiple comparisons)?

```
#Adding the inferential t-statistics to the df
all_subjects <- df %>% group_by(Subject) %>% summarise("Intercept" =
lm(Reaction~Days)$coefficients[1],
summary(lm(Reaction~Days))$coefficients[1,3],
summary(lm(Reaction~Days))$coefficients[1,4],
lm(Reaction~Days)$coefficients[2],
summary(lm(Reaction~Days))$coefficients[2,3],
summary(lm(Reaction~Days))$coefficients[2,4],
lm(Reaction~Days)$df.residual)

"t-value_intercept" =
"p-value_intercept" =
"Slope" =
"t-value_slope" =
"p-value_slope" =
"df" =
```

#Adding a collumn with marked "" if p<0.05.*

```
all_subjects$significant <- ifelse(all_subjects$p-value_slope<0.05,"*"," ")
```

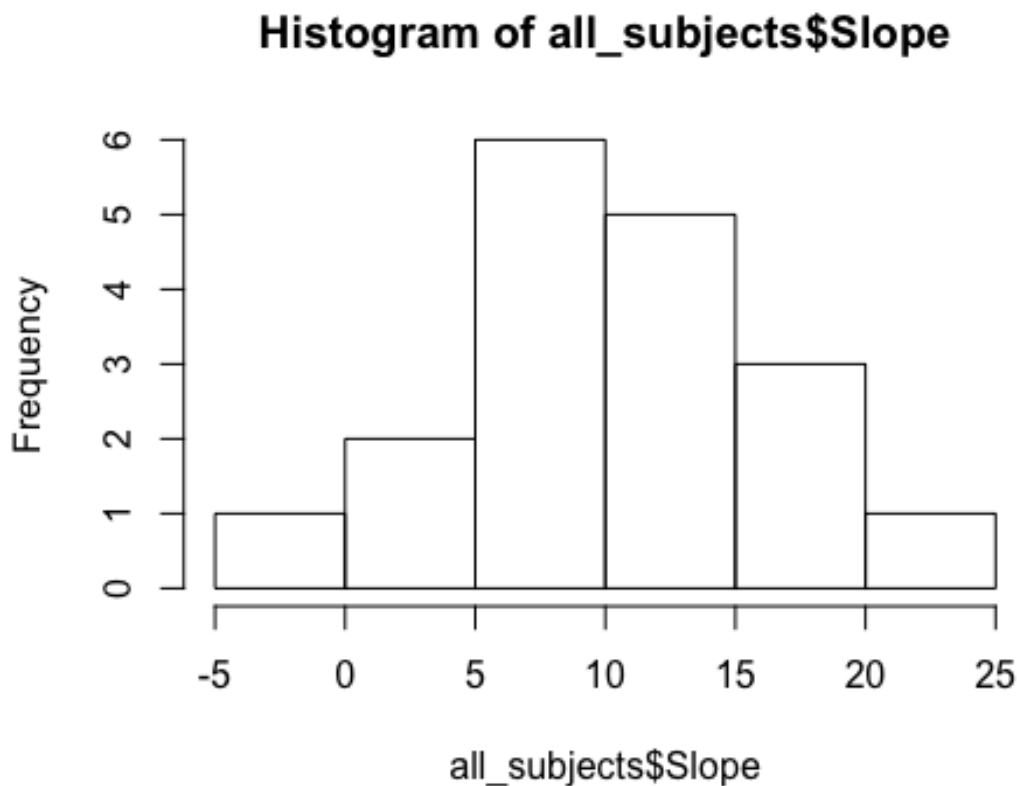
```
print.data.frame(all_subjects)
```

##	Subject	Intercept	t-value_intercept	p-value_intercept	Slope
## 1	308	244.1927	8.695486	2.385022e-05	21.764702
## 2	309	205.0549	39.311440	1.927496e-10	2.261785
## 3	310	203.4842	28.100452	2.778115e-09	6.114899
## 4	330	289.6851	22.105427	1.852824e-08	3.008073
## 5	331	285.7390	20.749421	3.050571e-08	5.266019
## 6	332	264.2516	7.382976	7.744761e-05	9.566768
## 7	333	275.0191	37.557051	2.772320e-10	9.142045
## 8	334	240.1629	19.877588	4.275274e-08	12.253141
## 9	335	263.0347	39.295935	1.933560e-10	-2.881034
## 10	337	290.1041	30.242212	1.551288e-09	19.025974
## 11	349	215.1118	26.118064	4.959747e-09	13.493933
## 12	350	225.8346	15.772768	2.608897e-07	19.504017
## 13	351	261.1470	19.523721	4.923090e-08	6.433498
## 14	352	276.3721	18.426353	7.749702e-08	13.566549
## 15	369	254.9681	27.400119	3.393323e-09	11.348109
## 16	370	210.4491	14.845934	4.174433e-07	18.056151
## 17	371	253.6360	17.211684	1.321176e-07	9.188445
## 18	372	267.0448	40.264853	1.592623e-10	11.298073
##	t-value_slope	p-value_slope	df	significant	
## 1	4.137485	3.264657e-03	8	*	
## 2	2.314848	4.931443e-02	8	*	
## 3	4.508107	1.980757e-03	8	*	
## 4	1.225416	2.552687e-01	8		
## 5	2.041462	7.550229e-02	8		
## 6	1.426926	1.914426e-01	8		
## 7	6.664912	1.583426e-04	8	*	
## 8	5.414117	6.352350e-04	8	*	
## 9	-2.297764	5.064731e-02	8		
## 10	10.588367	5.530467e-06	8	*	
## 11	8.746570	2.285006e-05	8	*	
## 12	7.272169	8.617903e-05	8	*	
## 13	2.567717	3.324544e-02	8	*	
## 14	4.828776	1.306668e-03	8	*	
## 15	6.510472	1.860407e-04	8	*	
## 16	6.799987	1.378251e-04	8	*	
## 17	3.328717	1.040424e-02	8	*	
## 18	9.094290	1.716323e-05	8	*	

Task 3: Across participants:

3.a: Use the slopes you found for each participant in exercise 2 as a new dataset. Test the hypothesis that the slopes are larger than zero against the null-hypothesis that the slopes are zero (i.e. no differences in response time exist as a function of time).

```
#Testing for normality  
hist(all_subjects$Slope)
```



#By visual inspection the data approximates normal distribution even though the small sample size of n=18.

#Making a 1-sample t-test to test if the slopes significantly differs from 0.

```
m <- t.test(all_subjects$Slope)  
m  
  
##  
## One Sample t-test  
##  
## data: all_subjects$Slope  
## t = 6.7715, df = 17, p-value = 3.264e-06
```



```
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
##    7.205956 13.728615
## sample estimates:
## mean of x
##    10.46729
```

3.b: Justify your use of test statistics.

T-test is used to test whether two means are statically significant different from each other. We used a 1-sample t-test of the slopes to test if they differed from 0, and in that way if there is a significant of sleep deprivation on reaction time.

3.c: Report inferential statistics.

Using an 1-sample t-test, we found that the mean gradient of the slopes significantly differed from 0, $t(17) = 6.77$, $p < 0.001$, (M slope = 10.47, M = 0), which indicates that sleep deprivation significantly modulates reaction time.

3.d: Make a plot with the mean reaction time and standard error bars for each day across participants and plot the averaged regression line in the same figure.

```
#Making a plot
ggplot(df, aes(Days, Reaction))+
  stat_summary(fun.y = mean, geom = "point", colour = "black")+
  stat_summary(fun.data = mean_se, geom = "errorbar", width = 0.3)+
  geom_smooth(method = lm, alpha = 0)+
  ggtitle("Mean reaction time pr. day across participants with se-bars")
```

Mean reaction time pr. day across participants with se-l

